

Abstract

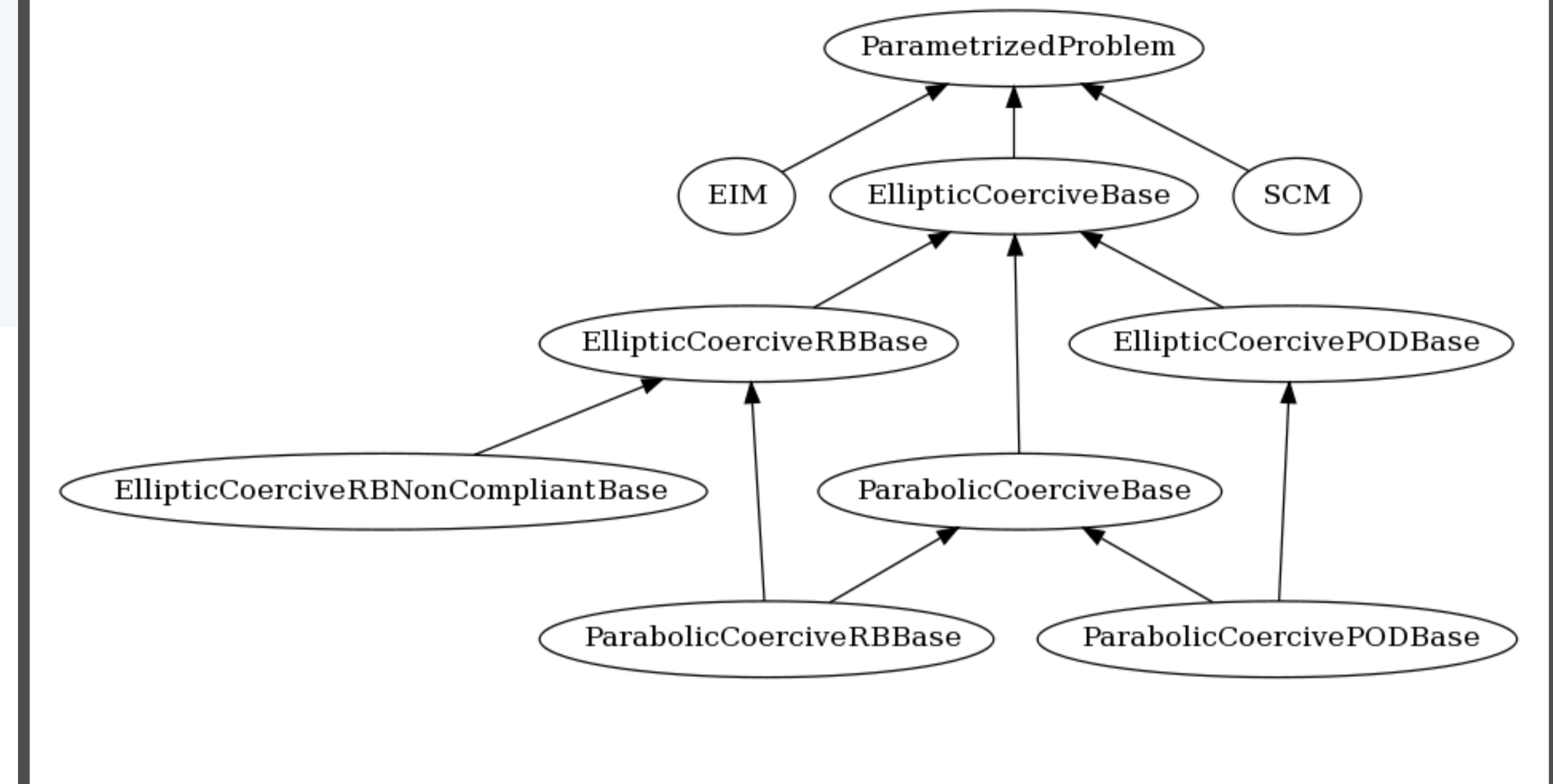
RBniCS [1] is a python-based library, developed on top of FEniCS [3], aimed at the development of reduced order models in the FEniCS environment. In particular, reduced order techniques such as the certified reduced basis method and proper orthogonal decomposition-Galerkin methods are implemented. The FEniCS project allows RBniCS to take advantage of the high-level (e.g., hu-

man readable) code used for the automated solution of partial differential equations. Thanks to the features of FEniCS the final user needs to prepare a short code to carry out a reduced order simulation.

It is ideally suited for novice users willing to learn reduced basis methods and reduced order modelling, thanks to an object-oriented approach and an intuitive and versatile python in-

terface. Indeed, it is a companion of the introductory reduced basis handbook [2], and has been already used in doctoral classes within the “Mathematical Analysis, Modelling, and Applications” PhD course at SISSA, as well as for courses within the “Master in High Performance Computing” jointly offered by SISSA and International Centre for Theoretical Physics (ICTP).

Object-oriented approach

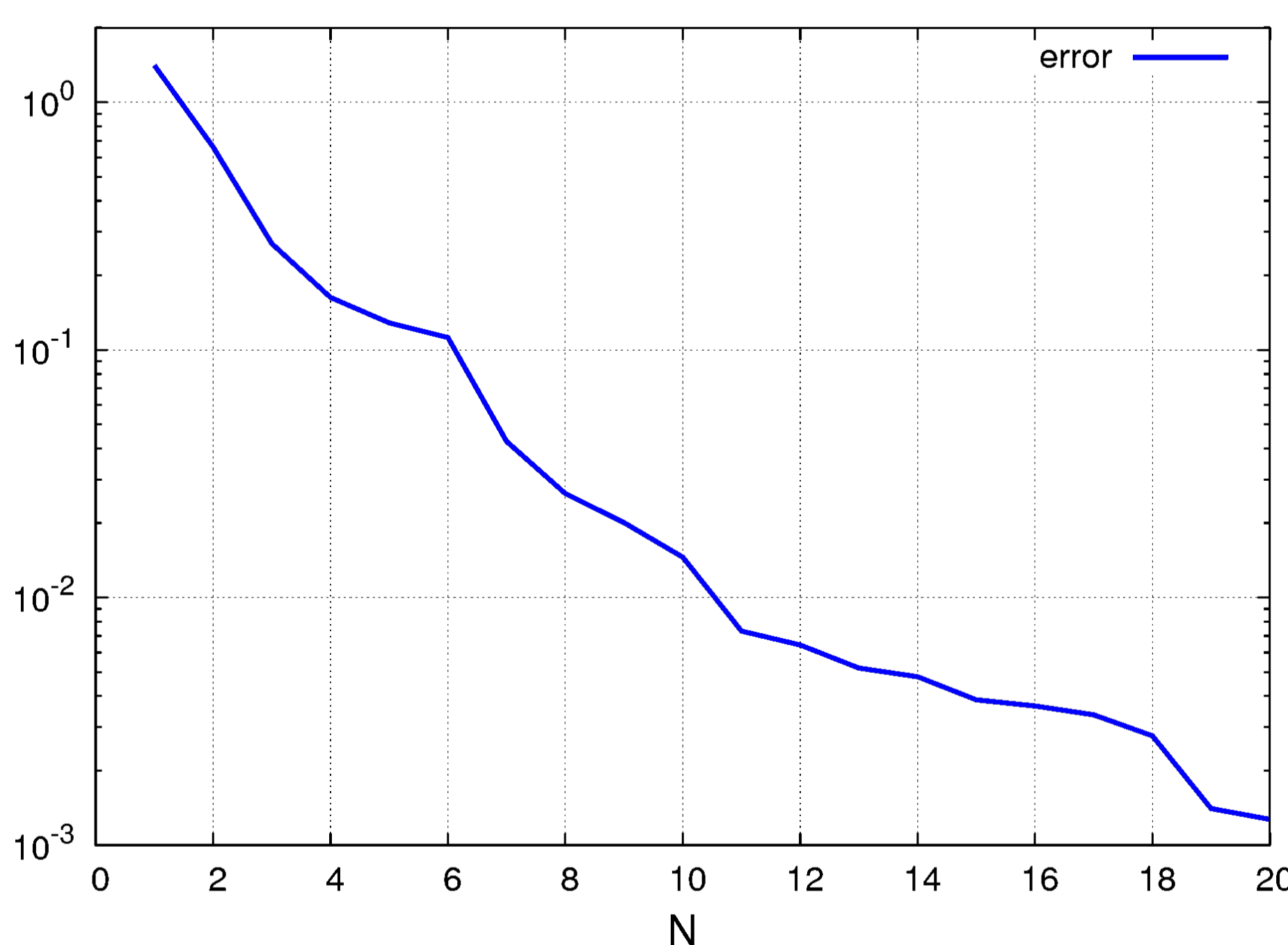


Template of a main program

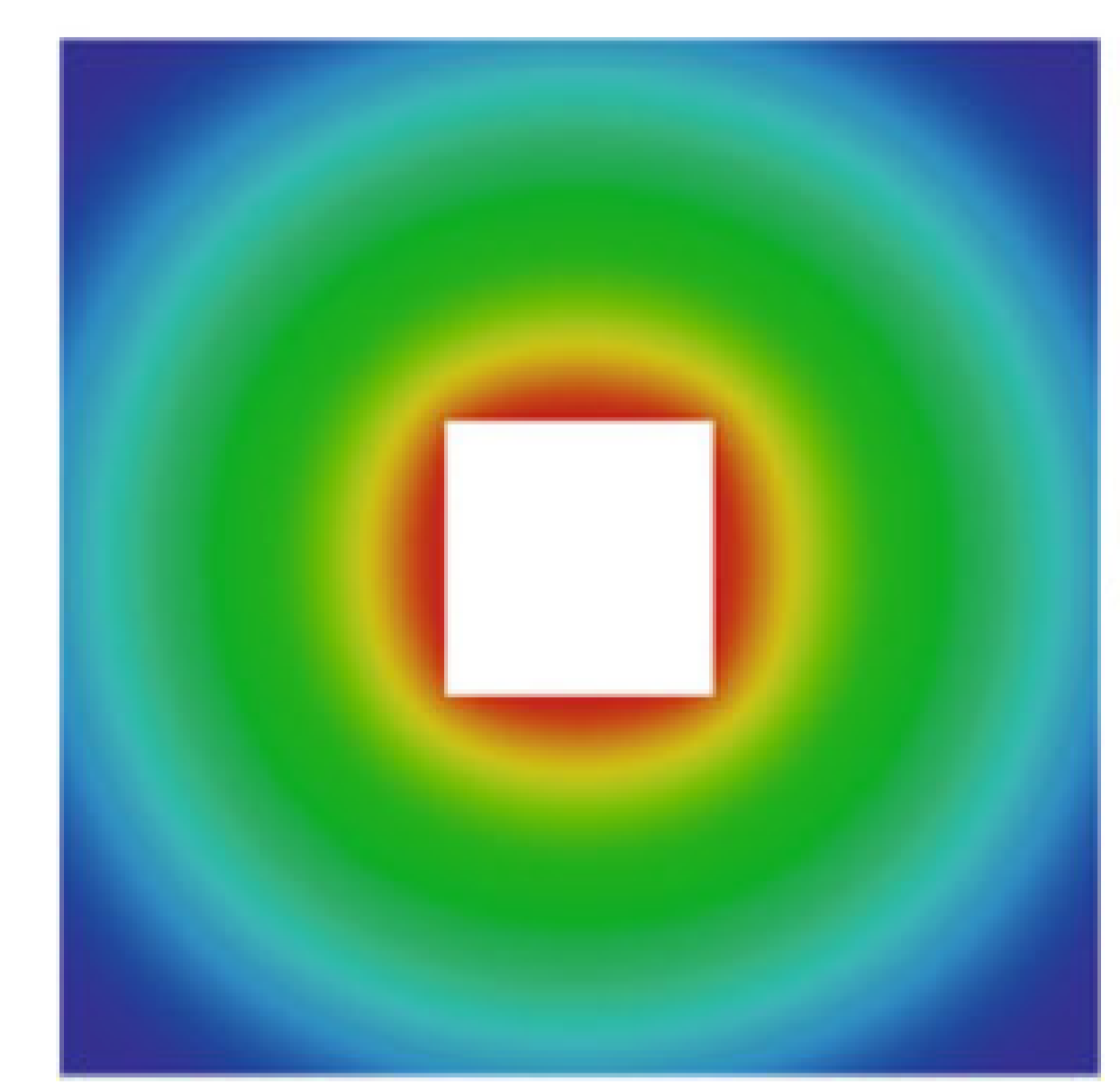
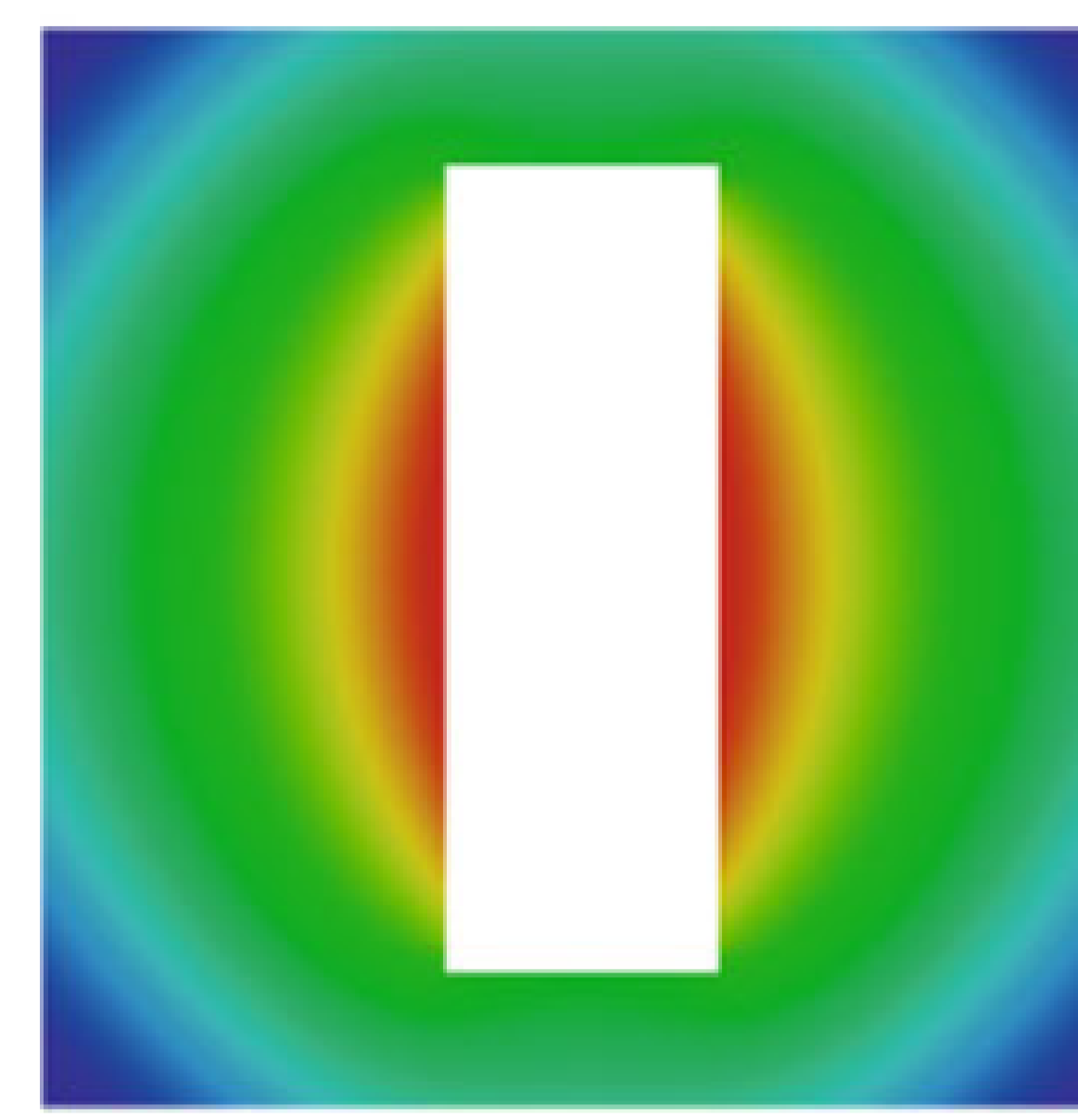
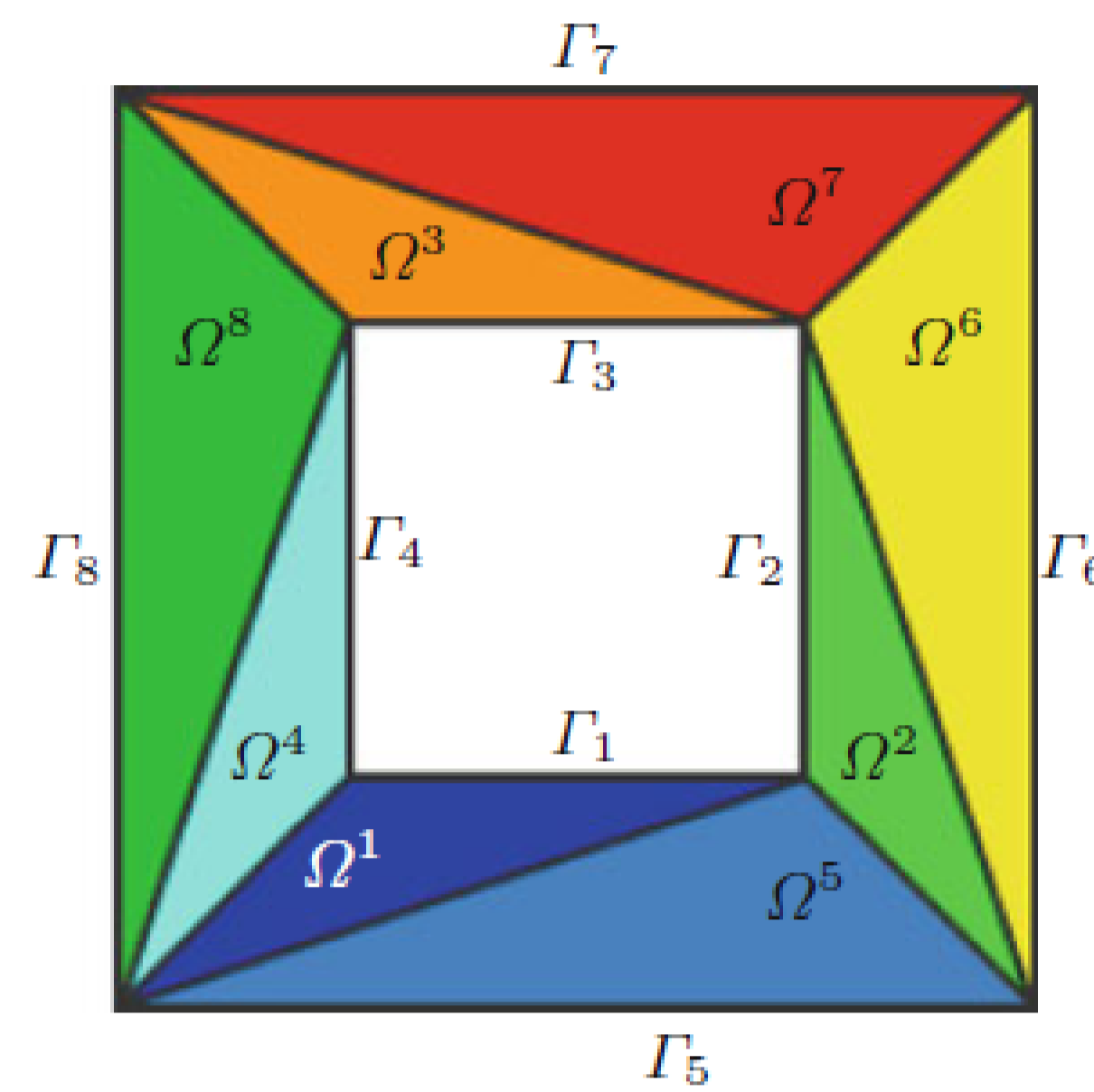
```
1 from dolfin import *
2 from RBniCS import *
3 mesh = Mesh("mesh.xml")
4 subd = MeshFunction("size_t", mesh,
5 "mesh_physical_region.xml")
6 bound = MeshFunction("size_t", mesh,
7 "mesh_facet_region.xml")
8 V = FunctionSpace(mesh, "Lagrange", 1)
9 tb = MyProblem(V, subd, bound)
10 mu_range = [(0.1, 1.0), (1.0, 10.0)]
11 tb.setmu_range(mu_range)
12 tb.setxi_train(1000)
13 tb.setNmax(25)
14 tb.offline()
15 online_mu = (0.5, 6.0)
16 tb.setmu(online_mu)
17 tb.online_solve()
18 tb.setxi_test(500)
19 tb.error_analysis()
```

Tutorials 1 and 2: basics

```
1 class MyProblem(EllipticCoerciveRBBBase):
2 def __init__(self, V, subd, bound):
3 ... # constructor ...
4 def compute_theta_a(self):
5 mu1 = self.mu[0]
6 return (mu1, 1.)
7 def compute_theta_f(self):
8 mu2 = self.mu[1]
9 return (mu2,)
10 def assemble_truth_a(self):
11 u = self.u; v = self.v; dx = self.dx
12 # Assemble
13 a0 = inner(grad(u), grad(v))*dx(1)
14 A0 = assemble(a0)
15 a1 = inner(grad(u), grad(v))*dx(2)
16 A1 = assemble(a1)
17 # Return
18 return (A0, A1)
19 def assemble_truth_f(self):
20 v = self.v; ds = self.ds
21 # Assemble
22 f0 = v*ds(1); F0 = assemble(f0)
23 # Return
24 return (F0,)
25 def get_alpha_lb(self):
26 return min(self.compute_theta_a())
```

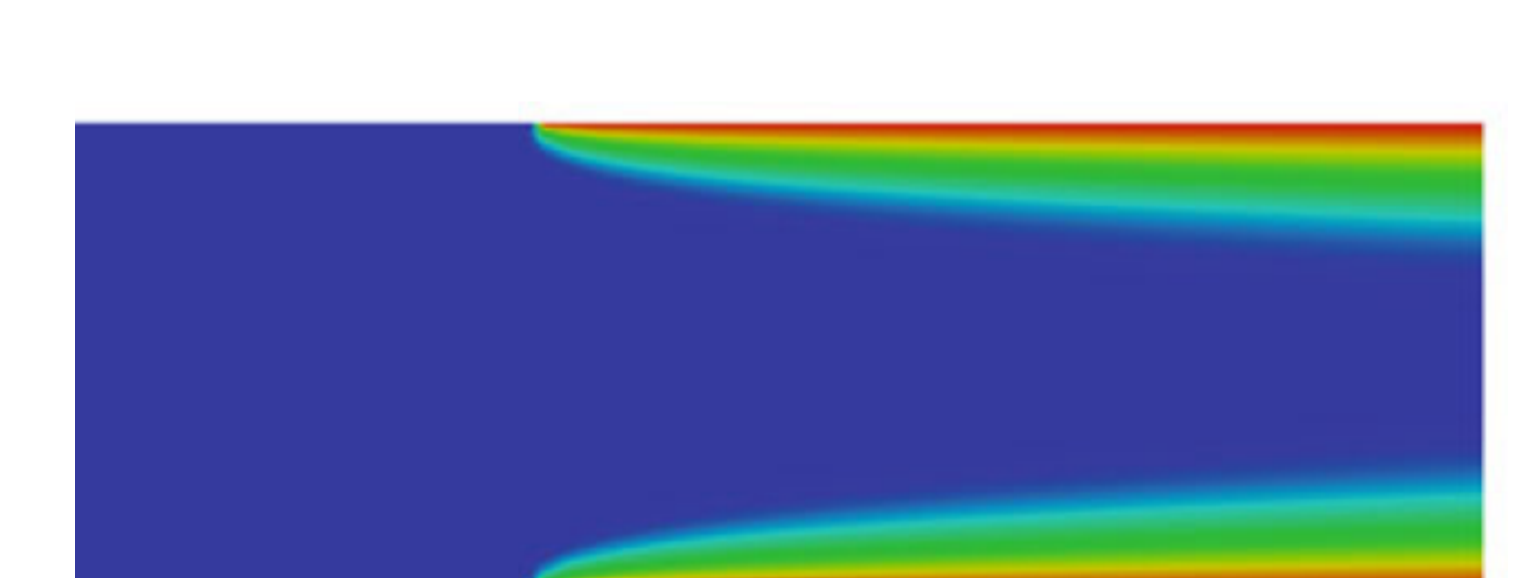
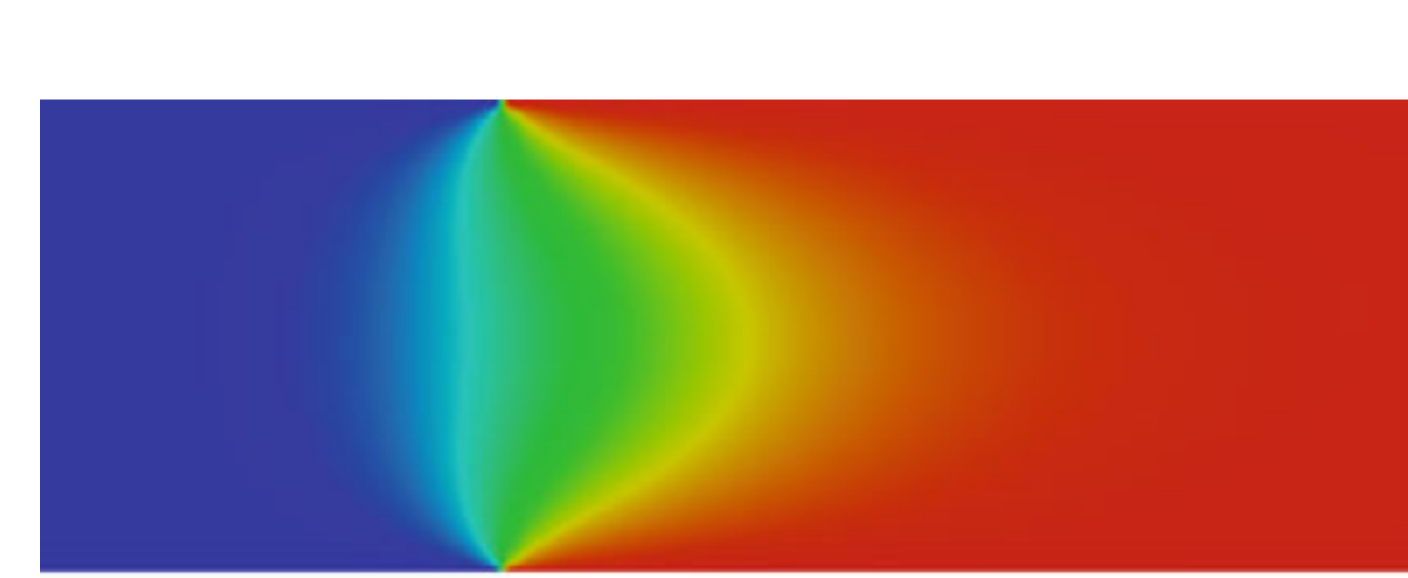
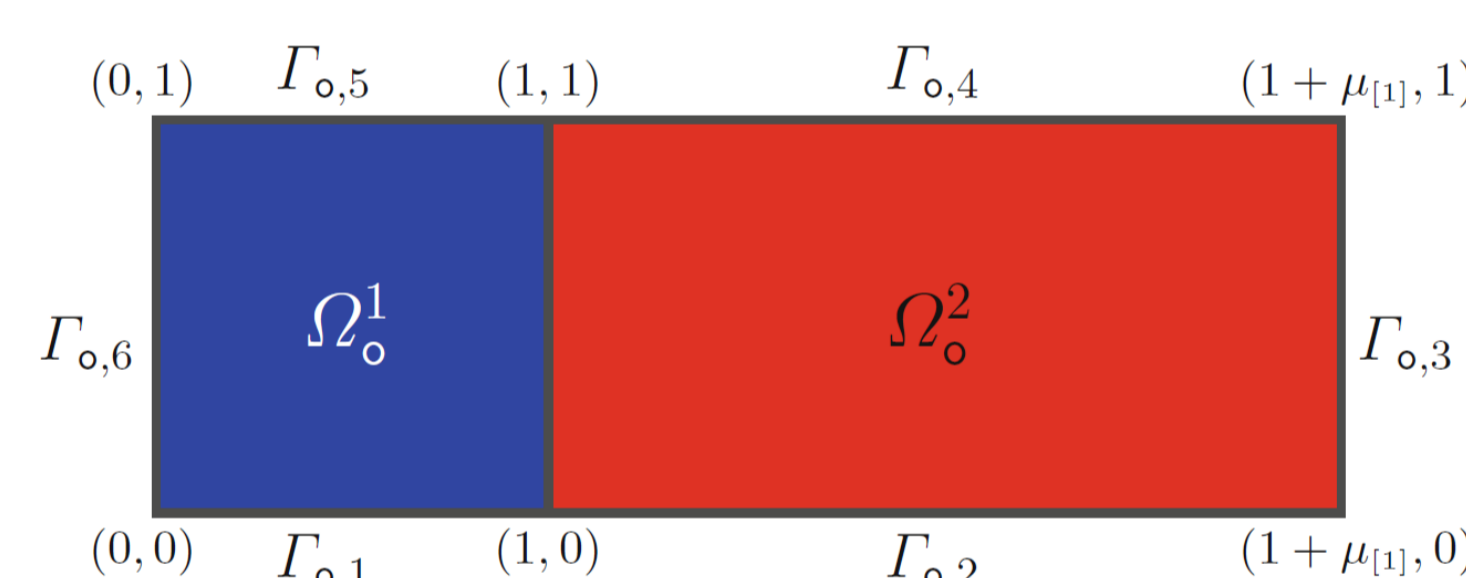


Tutorials 3 and 4: geometrical parametrization



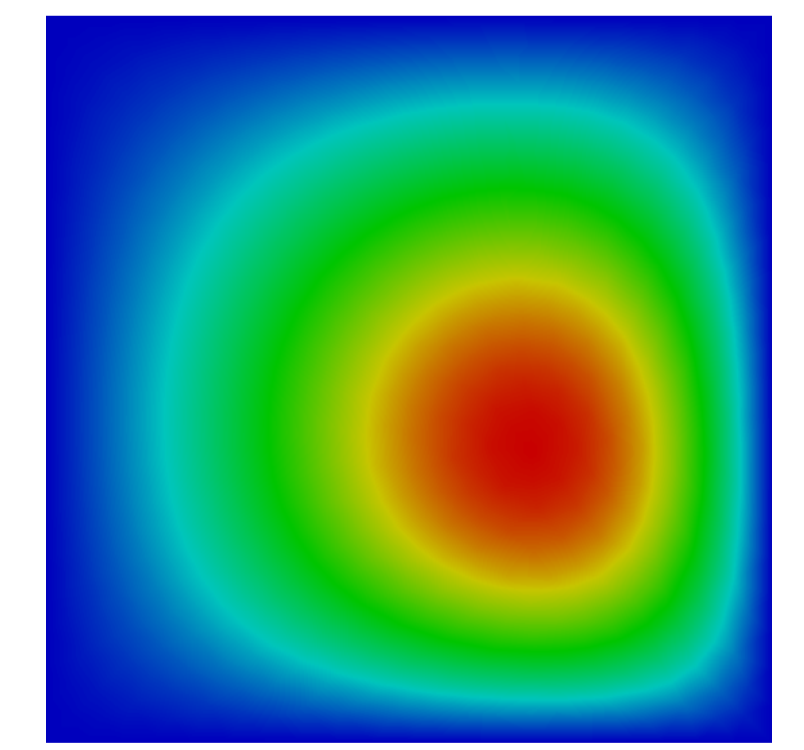
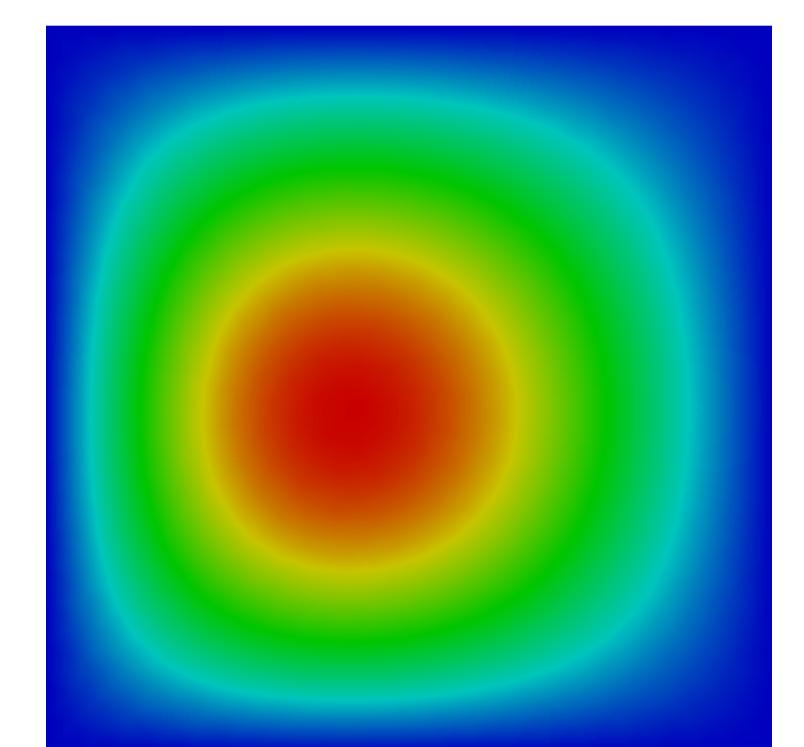
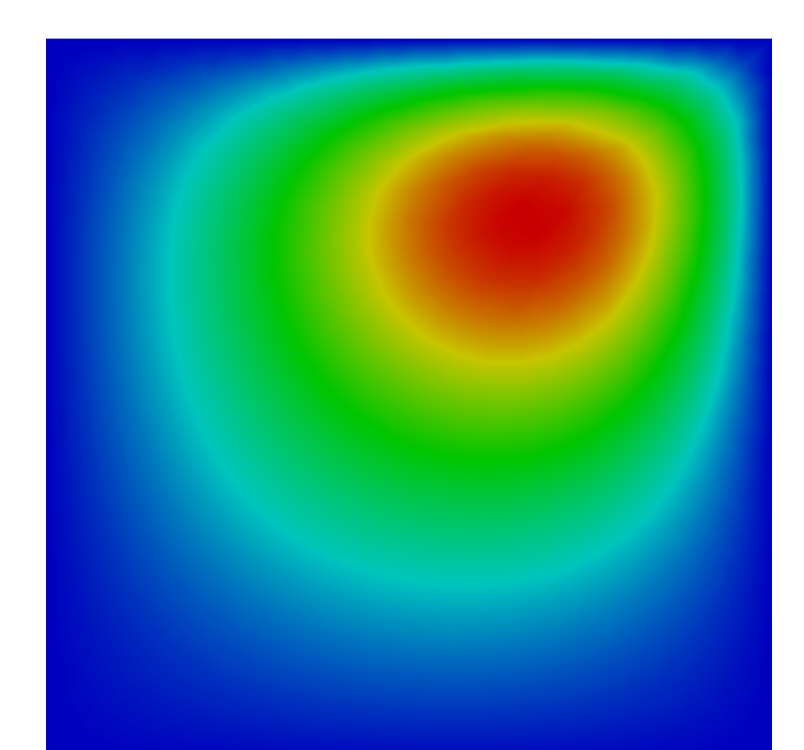
Tutorial 4: stability factor by SCM

```
1 class MyProblemWithSCM(EllipticCoerciveRBBBase):
2 def __init__(self, V, subd, bound):
3 ...
4 # Initialize an SCM object
5 self.SCM_obj = SCM(self)
6 def offline(self):
7 # Perform first the SCM offline phase, ...
8 bak_first_mu = tuple(list(self.mu))
9 self.SCM_obj.offline()
10 # ..., and then call the parent method.
11 self.setmu(bak_first_mu)
12 EllipticCoerciveRBBBase.offline(self)
13 def get_alpha_lb(self):
14 return self.SCM_obj.get_alpha_LB(self.mu)
```



Tutorial 5: affinity assumption recovery by EIM

```
1 class MyProblemWithEIM(EllipticCoerciveRBBBase):
2 def __init__(self, V, subd, bound):
3 ...
4 # Initialize an EIM object
5 self.EIM_obj = EIM(self)
6 self.EIM_obj.parametrized_function = "exp(-2*pow(x[0]-mu_1, 2)
7 -2*pow(x[1]-mu_2, 2))"
8 def compute_theta_f(self):
9 self.EIM_obj.setmu(self.mu)
10 return self.EIM_obj.compute_interpolated_theta()
11 def assemble_truth_f(self):
12 v = self.v; dx = self.dx
13 # Call EIM
14 self.EIM_obj.setmu(self.mu)
15 interpolated_gaussian = self.EIM_obj.
16 assemble_mu_independent_interpolated_function()
17 # Assemble
18 all_F = ()
19 for q in range(len(interpolated_gaussian)):
20 f_q = interpolated_gaussian[q]*v*dx
21 all_F += (assemble(f_q),)
22 # Return
23 return all_F
```



How to get RBniCS

RBniCS is freely available under the GNU LGPL, version 3, at



<http://mathlab.sissa.it/rbnics>

References

- [1] F. Ballarin, A. Sartori, and G. Rozza. RBniCS – reduced order modelling in FEniCS. <http://mathlab.sissa.it/rbnics>, 2015.
- [2] J. S. Hesthaven, G. Rozza, and B. Stamm. *Certified Reduced Basis Methods for Parametrized Partial Differential Equations*. SpringerBriefs in Mathematics. Springer, 2015.
- [3] A. Logg, K.-A. Mardal, G. N. Wells, et al. *Automated Solution of Differential Equations by the Finite Element Method*. Springer, 2012.

Acknowledgements

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